

Visual object detection system

Abstract:-

This study proposes a deep neural network architecture for the phone location detection model. Initially multiple images have been trained using VGG16 Architecture , a phone is considered to be detected correctly on a test image if the final object coordinates of the test image is within a radius of 0.05 (normalized distance) centered on the phone. Perfect detection performance is not the main goal of this test. For this prototype, your algorithm is expected to detect a phone correctly on 4 out of the 8 test images and to detect at least 70% correctly on the provided labeled dataset. If you do not have enough time, please focus on a submission with clean, well structured code, rather than on the perfect performance.

Input data details:-

| | |
|------------------|--------|
| Parameters | |
| Images | 120 |
| Number of epochs | 200 |
| Optimizer | Adam |
| Learning rate | 1e - 4 |
| batch size | 16 |
| Layers | 5 |

Bounding box regression model using VGG net is used for object detection

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|---|--|
| <pre>bboxHead = Dense(128, activation="relu")(flatten) bboxHead = Dense(64, activation="relu")(bboxHead) bboxHead = Dense(32, activation="relu")(bboxHead) bboxHead = Dense(16, activation="relu")(bboxHead) bboxHead = Dense(8, activation="relu")(bboxHead) bboxHead = Dense(2, activation="sigmoid")(bboxHead)</pre> | |
| Train_phone_finder.py | This file will take the folder path which contains labels and images |
| Find_phone.py | It will take input image to be tested |

Conclusion:-

The model worked with 70% accuracy with a test image of 90:10 split. This python script will be responsible for:

1. Loading our phone image training data from disk (i.e., both class labels and center coordinates)
2. Loading VGG16 (pre-trained on ImageNet), removing the fully-connected classification layer head from the network, and inserting our bounding box regression layer head
3. Fine-tuning the regression layer head on our training data